

```

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.preprocessing import StandardScaler
from sklearn.cluster import KMeans
from scipy.cluster.hierarchy import linkage, dendrogram, fcluster

# Load the dataset
data = pd.read_csv("/content/sales_data_sample.csv",
encoding='latin1')
data.head()

{"type": "dataframe", "variable_name": "data"}
```

Select numeric columns for clustering

```
num_data = data.select_dtypes(include=['float64', 'int64']).copy()
```

Handle missing values if any

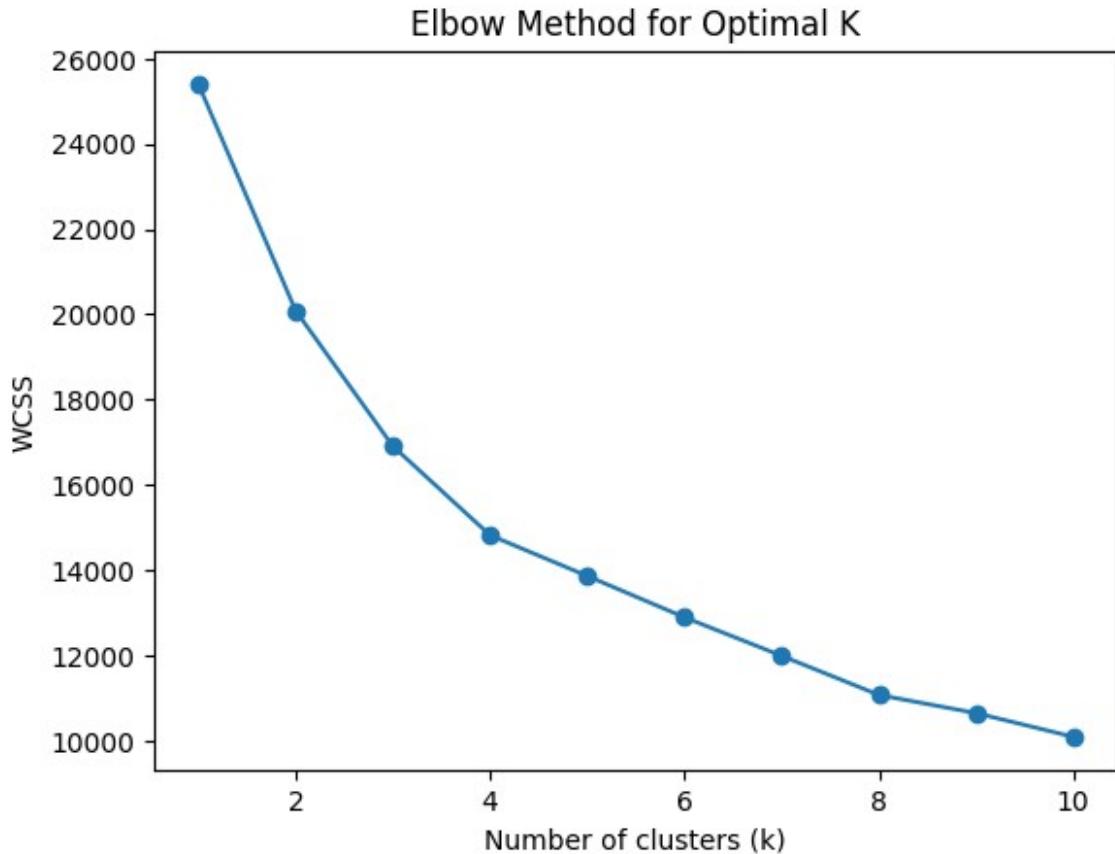
```
num_data = num_data.dropna()
```

Standardize the data

```
scaler = StandardScaler()
scaled_data = scaler.fit_transform(num_data)
```

```
wCSS = []
for i in range(1, 11):
    kmeans = KMeans(n_clusters=i, random_state=42)
    kmeans.fit(scaled_data)
    wCSS.append(kmeans.inertia_)

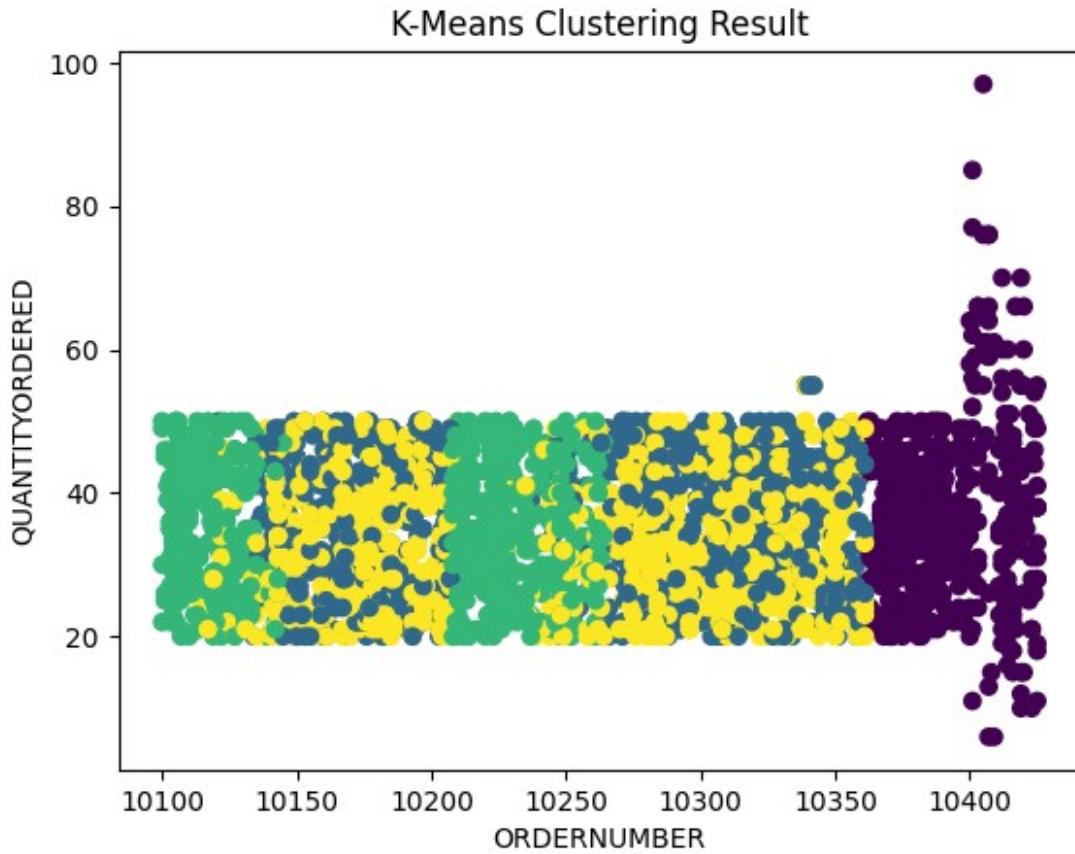
# Plot the elbow curve
plt.plot(range(1, 11), wCSS, marker='o')
plt.title('Elbow Method for Optimal K')
plt.xlabel('Number of clusters (k)')
plt.ylabel('WCSS')
plt.show()
```



```
# Choose k based on elbow result (for example, k=4)
kmeans = KMeans(n_clusters=4, random_state=42)
kmeans_labels = kmeans.fit_predict(scaled_data)

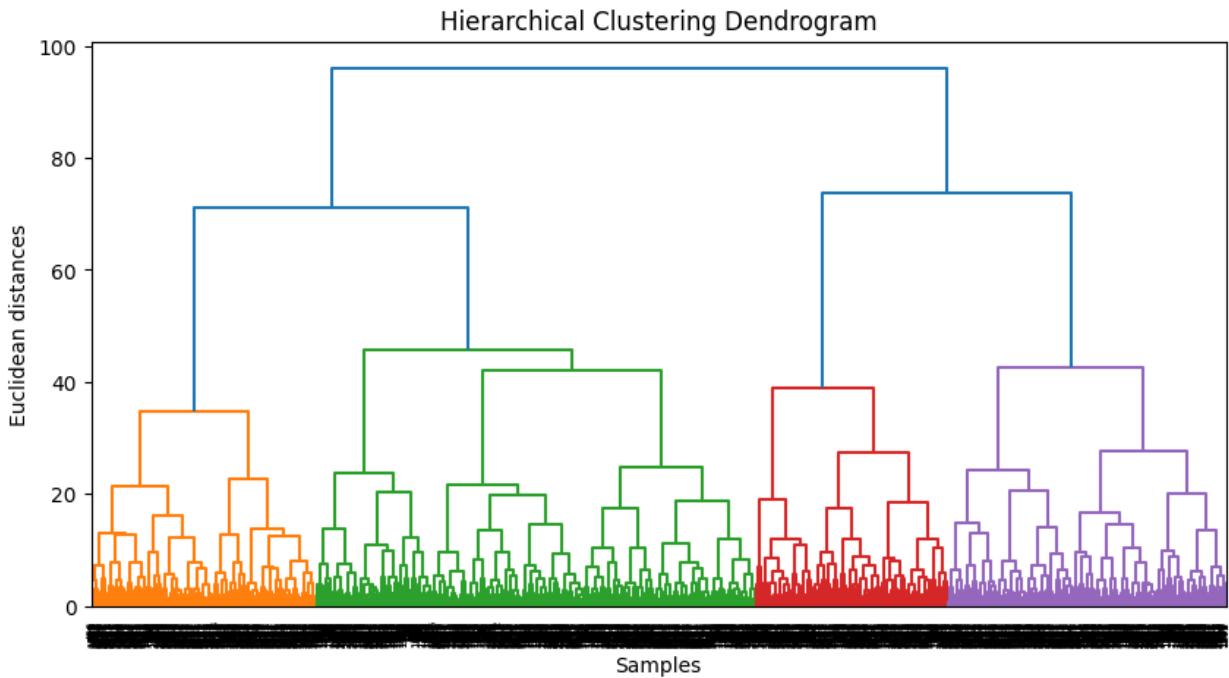
# Add cluster labels to dataset
num_data['Cluster'] = kmeans_labels

# Visualize clusters using two features
plt.scatter(num_data.iloc[:, 0], num_data.iloc[:, 1], c=kmeans_labels,
cmap='viridis')
plt.title('K-Means Clustering Result')
plt.xlabel(num_data.columns[0])
plt.ylabel(num_data.columns[1])
plt.show()
```



```
# Perform hierarchical clustering using Ward's method
Z = linkage(scaled_data, method='ward')

# Plot dendrogram
plt.figure(figsize=(10, 5))
dendrogram(Z)
plt.title('Hierarchical Clustering Dendrogram')
plt.xlabel('Samples')
plt.ylabel('Euclidean distances')
plt.show()
```



```
# Suppose we choose 4 clusters based on the dendrogram
hier_labels = fcluster(Z, t=4, criterion='maxclust')
num_data['HCluster'] = hier_labels

# Visualize first two features
plt.scatter(num_data.iloc[:, 0], num_data.iloc[:, 1], c=hier_labels,
cmap='plasma')
plt.title('Hierarchical Clustering Result')
plt.xlabel(num_data.columns[0])
plt.ylabel(num_data.columns[1])
plt.show()
```

Hierarchical Clustering Result

